# LAB 3

### STAT 28

### February 02, 2017

Welcome to the lab 3! In this lab, you will

- 1) plot your density curves and violin plot;
- 2) obtain some basic statistics by groups;
- 3) implement a permutation test;

We will continue to use the rent price dataset from the lab 2. In the table **craigslist.csv**, each posting record (row) contains the following information:

- time: posting time
- price: apartment/housing monthly rent price
- size: apartment/housing size (ft^2)
- brs: number of bedrooms
- title: posting title
- link: posting link, add "https://sfbay.craigslist.org" to visit the posting page
- location: cities

Read in data. Create one.bedrooms data frame.

## Install packages

R is very powerful in statistical analysis mainly due to a huge community that supports it. Experts contribute to R through packages which are easily accessible (through CRAN). There are two ways to install R packages.

#### Installation using Studio Interface

- Open you RStudio.
- Click Packages window in the bottom right panel and then click install.
- A window named install packages will pop up. Enter the name of packages you want to install. For example, vioplot from lecture 1. Make sure you checked install dependencies and then click Install.
- If you see the messages in the console similar to the following, you've successfully installed the package! Sometimes the messages will be much longer because many R packages use the code from others (dependencies), and R will need to download and install the dependencies as well.

```
> install.packages("vioplot")
Installing package into '/home/jovyan/R/x86_64-pc-linux-gnu-library/3.3'
(as 'lib' is unspecified)
trying URL 'https://mran.revolutionanalytics.com/snapshot/2017-01-16/src/contrib/vioplot_
0.2.tar.gz'
Content type 'unknown' length 3801 bytes
downloaded 3801 bytes
* installing *source* package 'vioplot' ...
** preparing package for lazy loading
** help
*** installing help indices
** building package indices
** testing if installed package can be loaded
* DONE (vioplot)
The downloaded source packages are in
        '/tmp/RtmphoSPbh/downloaded_packages'
```

• The package name will then appear in the list of Packages window. There are already a collection of packages in the list, which we previously installed for you.

#### Installation using R Code

There is a much quicker alternative than clicking bottoms in the first method. You will only need to run the following code:

```
install.packages("vioplot")

## Installing package into '/Users/jorothygong/Library/R/3.3/library'
## (as 'lib' is unspecified)

##

## The downloaded binary packages are in
## /var/folders/1p/phtqfgjn1332c0x1trzm7n800000gn/T//RtmpNYSI79/downloaded_packages

Loading Packages
```

To use functions from your installed packages, you will need to load them by running library function. For example, to load the vioplot package:

```
library(vioplot)

## Loading required package: sm

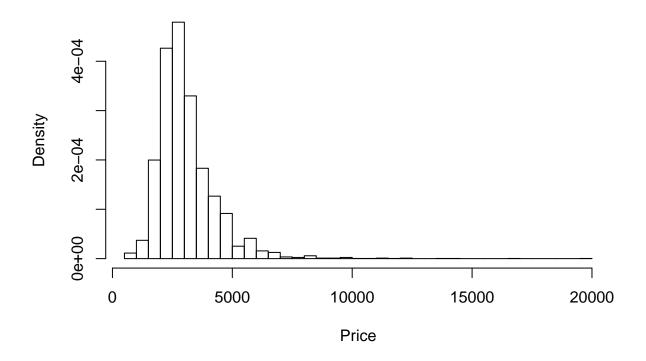
## Package 'sm', version 2.2-5.4: type help(sm) for summary information
```

## Density Curves and Violin Plot

```
Recall the histogram you plotted with freq = FALSE in lab 2:
```

```
hist(craigslist$price, xlab = "Price", freq = FALSE, breaks = 50,
    main = "Histogram of montly rent")
```

## Histogram of montly rent



#### Exercise 1

Now fit a density curve for the monthly rent on top of the histogram.

HINT1: Use the function density and refer to Example of Flight data in section 4.3.1 (Kernel density estimation) of Lecture 1.

HINT2: You will need to first draw a histogram, and then plot the density curve with argument add = TRUE in the following chunk.

```
# insert your code here
```

#### Exercise 2

Draw the violin plot of one bedroom rent price by cities. (Follow the example given in lectures (section 4.4.3), do not forget to add titles and legend)

```
library(vioplot)
source("http://www.stat.berkeley.edu/~epurdom/RcodeForClasses/myvioplot.R")
# insert your code here
```

### Summarize dataset by groups

In this dataset, we are more interested in the summaries of rent price by cities. The tapply function is useful when we need to break a vector into groups (subvectors), and apply a function (for example, the mean function) within each group. The usage is:

tapply(Variable\_of\_interests, Factor\_vector\_representing\_groups, Function\_you\_want\_to\_apply\_on)

For example, to obtain the median rent price by cities.

## tapply(craigslist\$price, craigslist\$location, median)

##	alameda	albany / el cerrito	berkeley
##	2497.5	2300.0	2885.0
##	emeryville	menlo park	mountain view
##	3095.0	3600.0	2862.5
##	oakland	palo alto	redwood city
##	2495.0	3395.0	3152.0
##	richmond	sunnyvale	
##	2850.0	2625.0	

You can write and apply your own functions. For example, to get the percentage of rent price less than \$2000/m onth by city.

## tapply(craigslist\$price, craigslist\$location, function(x){mean(x < 2000)})

##	alameda	albany / el cerrito	berkeley
##	0.16500000	0.21621622	0.10101010
##	emeryville	menlo park	mountain view
##	0.03809524	0.01285347	0.08555133
##	oakland	palo alto	redwood city
##	0.27809308	0.03802817	0.06333973
##	richmond	sunnyvale	
##	0.18625277	0.08795812	

The rent price in Berkeley is much better than Palo Alto! The median monthly rent is much lower. And the percentage of rent price less than \$2000 per much is higher. But do not rush to conclusions, let us break down the dataset further more.

#### Exercise 3

Use tapply to get following statistics for each city.

(a) The percentage of one bedrooms;

```
# insert code here save the precentage of one bedrooms by cites as
# 'pct.1b'
```

(b) the median price of one bedrooms. (You use the subset one.bedrooms created above)

```
# insert code here save the median of one bedrooms by cites as
# 'med.ib'
```

There are more one-bedroom rent postings at Berkeley. The median prices of one-bedrooms are less different for Berkeley and Palo Alto. The fact that the overall median price differs may probability caused by the large proportion of small apartment postings at Berkeley. How you obtain the sample may greatly influence your results. If we look at a stratified sampling dataset, where houses/apartments with 1, 2, 3 bedrooms account for 40%, 40%, 20% of the total samples of each city.

```
prop = c(0.4, 0.4, 0.2)
samples = c()
for (city in unique(craigslist$location)){
   for (b in 1:3){
      samples = c(samples, sample(which(craigslist$brs == b & craigslist$location == city), prop[b]*60))
   }
}
craigslist.srs <- craigslist[samples, ]</pre>
```

Now we look at the median rent price by cities for the stratified sampling dataset.

tapply(craigslist.srs\$price, craigslist.srs\$location, median)

berkeley	lbany / el cerrito	alameda	##
2925.0	2300.0	2550.0	##
mountain view	menlo park	emeryville	##
2898.5	3325.0	3095.0	##
redwood city	palo alto	oakland	##
3018.5	3350.0	2397.5	##
	sunnyvale	richmond	##
	2772.5	2690.0	##

Below is the percentage of rent price less than \$2000/month by cities for the stratified sampling dataset.

tapply(craigslist.srsprice, craigslist.srscontinuous)(mean(x < 2000)})

```
##
                alameda albany / el cerrito
                                                         berkeley
##
            0.18333333
                                  0.28333333
                                                       0.03333333
##
            emeryville
                                  menlo park
                                                    mountain view
##
            0.05000000
                                  0.0000000
                                                       0.10000000
##
                oakland
                                   palo alto
                                                     redwood city
##
            0.31666667
                                  0.0666667
                                                       0.08333333
##
              richmond
                                   sunnyvale
##
            0.21666667
                                  0.10000000
```

Now the difference of price median between Berkeley and Palo Alto reduced compared to the SRS sample. This is a case where simple random samples may be misleading. And the results from stratified samples may well depend on the how you assign the proportions to each stratum. Care must be taken before you reach conclusions.

#### Permutation test

#### Exercise 4

You will do a permutation test to compare the average one bedroom apartment rent price in Berkeley and Palo Alto.

To repeat things in R, there is a more convenient and efficient way than using for loop. The function replicate is designed for evaluating an expression repeatedly. For example, to get a vector of length 1000 whose elements are generated from the sum of five normal distributed samples:

```
example <- replicate(1000, sum(rnorm(5)))</pre>
```

(a) Subset the dataset to only consider one one bedroom apartment in Berkeley and Palo Alto.

```
# insert code here save the data frame of one bedroom postings
# in Berkeley and Palo Alto as
# 'subset'
```

(b) Calculate the number of postings for Berkeley in the data frame subset.

```
# insert code here save the number of postings for Berkeley as
# 'no.berkeley'
```

(c) Calculate the observed statistics, i.e., the difference between the mean of Berkeley and Palo Alto one bedroom rent price.

```
# insert code here save the observed statistics as
# 'stat.obs'
```

(d) Calculate the permuted statistics, repeat for 1000 times. HINT: use sample function to sample from a group of observations. You can use either for loop or the replicate function introduced above.

```
# insert code here save the observed statistics as
# 'stat.bootstrap'
set.seed(20172828)
```

(e) Calculate the p-value.

```
# insert code here save the observed statistics as
# 'p.value'
```